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# Basic Search

Philipp Koehn

11 February 2025



# Problem Solving Agents



- Restricted form of general agent
- Approach
  - formulate goal
  - formulate problem
  - search for solution
- Solution is a sequence of actions

# problem types

# Problem Types



3

- **Deterministic, fully observable**  $\implies$  **single-state problem**
  - agent knows exactly which state it will be in
  - solution is a sequence■
- **Non-observable**  $\implies$  **conformant problem**
  - Agent may have no idea where it is
  - solution (if any) is a sequence■
- **Nondeterministic** and/or **partially observable**  $\implies$  **contingency problem**
  - percepts provide **new** information about current state
  - solution is a **contingent plan** or a **policy**
  - often **interleave** search, execution■
- **Unknown state space**  $\implies$  **exploration problem** (“online”)

# Example: Vacuum World



**Single-state**, start in #5. **Solution?**

[*Right, Suck*]

**Conformant**, start in {1, 2, 3, 4, 5, 6, 7, 8}

e.g., *Right* goes to {2, 4, 6, 8}. **Solution?**

[*Right, Suck, Left, Suck*]

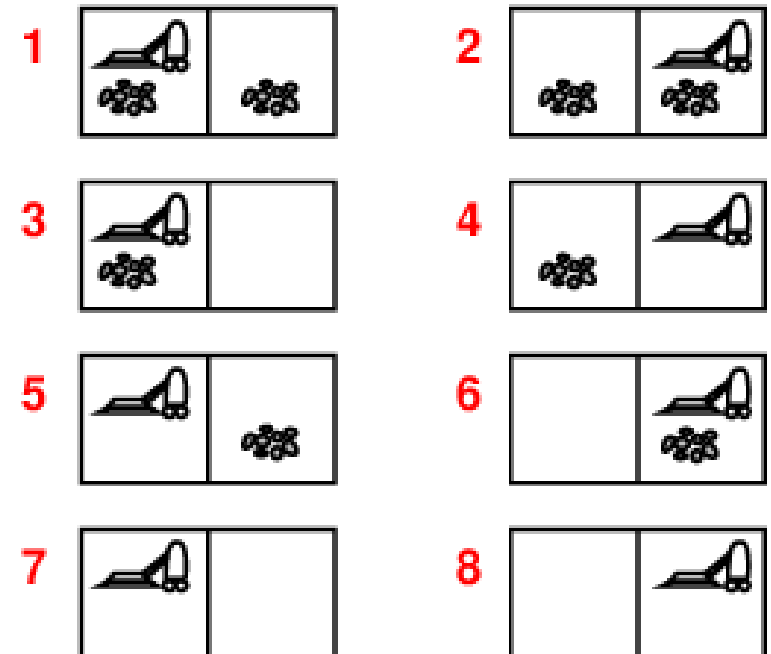
**Contingency**, start in #5

Murphy's Law: *Suck* can dirty a clean carpet

Local sensing: dirt, location only.

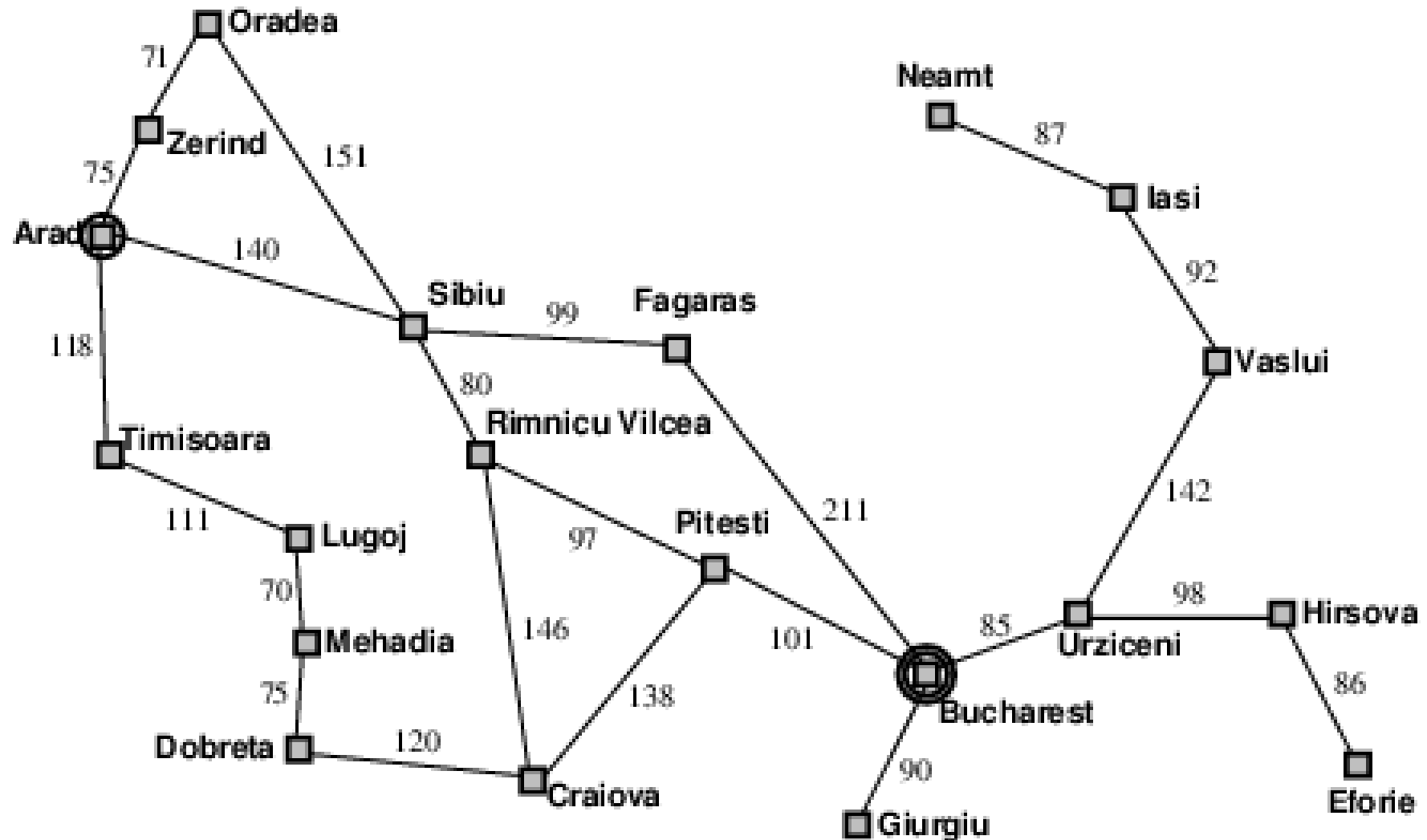
**Solution?**

[*Right, if dirt then Suck*]



# problem formulation

# Example: Romania



# Example: Romania



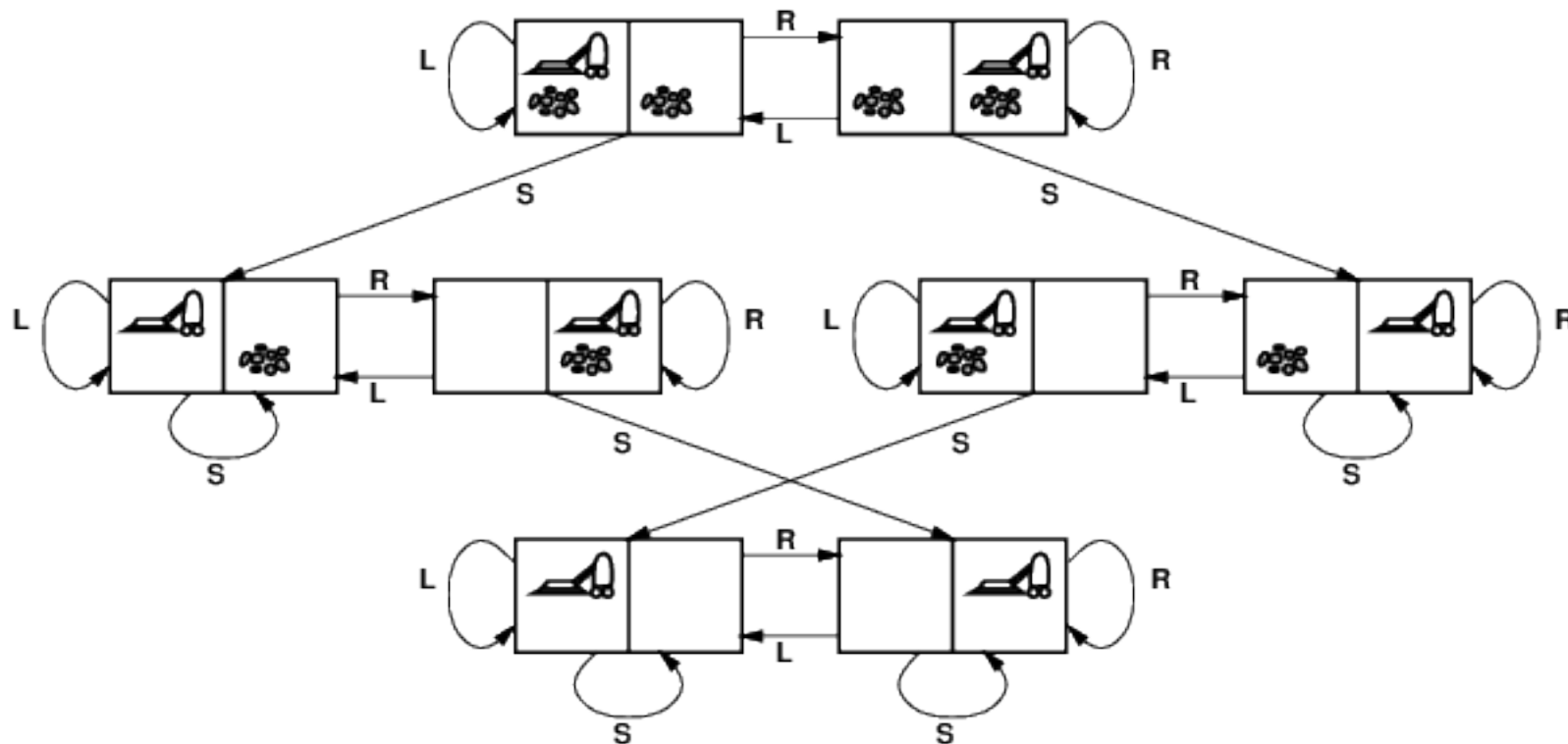
- On holiday in Romania; currently in Arad
- Flight leaves tomorrow from Bucharest
- Formulate goal
  - be in Bucharest
- Formulate problem
  - **states**: various cities
  - **actions**: drive between cities
- Find solution
  - sequence of cities, e.g., Arad, Sibiu, Fagaras, Bucharest

# Single-State Problem Formulation



- A **problem** is defined by four items:
  - **initial state** e.g., “at Arad”■
  - **successor function**  $S(x)$  = set of action–state pairs  
e.g.,  $S(\text{Arad}) = \{\langle \text{Arad} \rightarrow \text{Zerind}, \text{Zerind} \rangle, \dots\}$ ■
  - **goal test**, can be  
**explicit**, e.g.,  $x = \text{“at Bucharest”}$   
**implicit**, e.g.,  $\text{NoDirt}(x)$ ■
  - **path cost** (additive)  
e.g., sum of distances, number of actions executed, etc.  
 $c(x, a, y)$  is the **step cost**, assumed to be  $\geq 0$ ■
- A **solution** is a sequence of actions leading from the initial state to a goal state

# Example: Vacuum World State Space Graph



**states?**: integer dirt and robot locations (ignore dirt **amounts** etc.)

**actions?**: *Left, Right, Suck, NoOp*

**goal test?**: no dirt

**path cost?**: 1 per action (0 for *NoOp*)

# Example: The 8-Puzzle

7	2	4
5		6
8	3	1

Start State

1	2	3
4	5	6
7	8	

Goal State

**states?**: integer locations of tiles

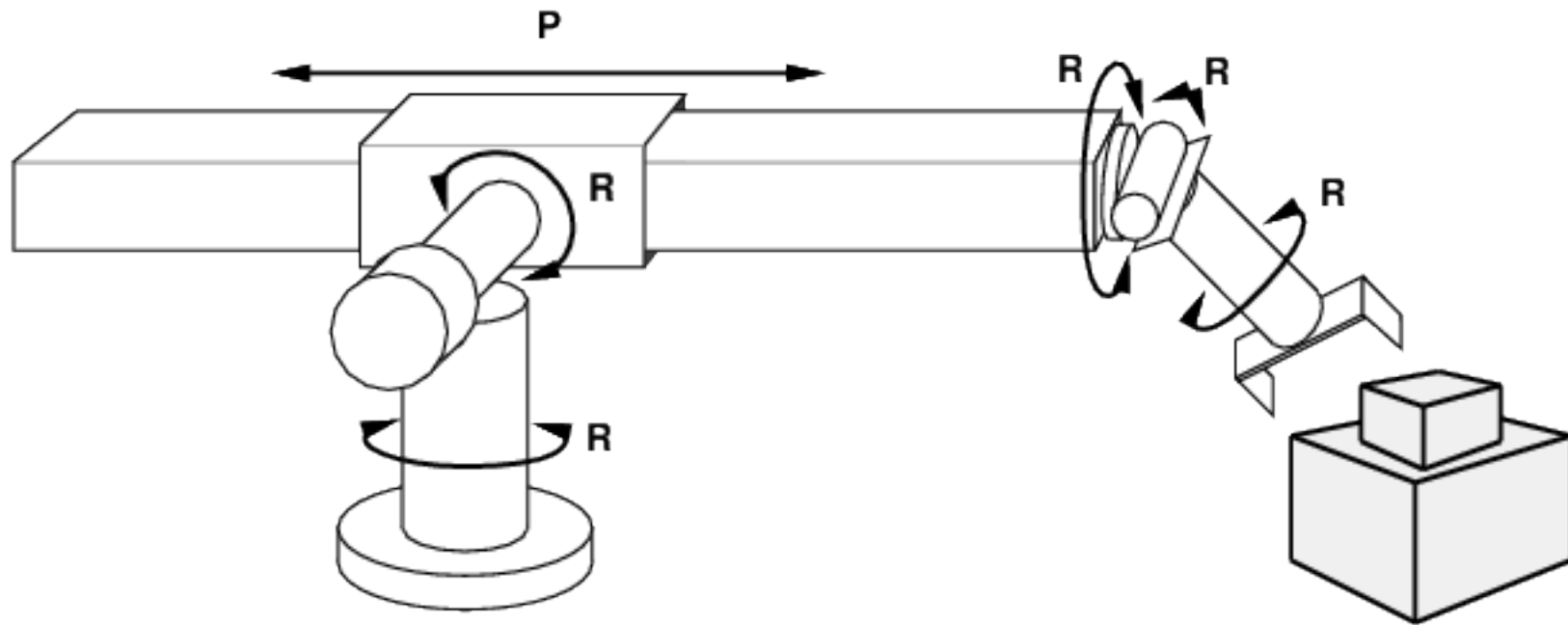
**actions?**: move blank left, right, up, down

**goal test?**: = goal state (given)

**path cost?**: 1 per move

[Note: optimal solution of  $n$ -Puzzle family is NP-hard]

# Example: Robotic Assembly



**states?**: real-valued coordinates of robot joint angles

parts of the object to be assembled

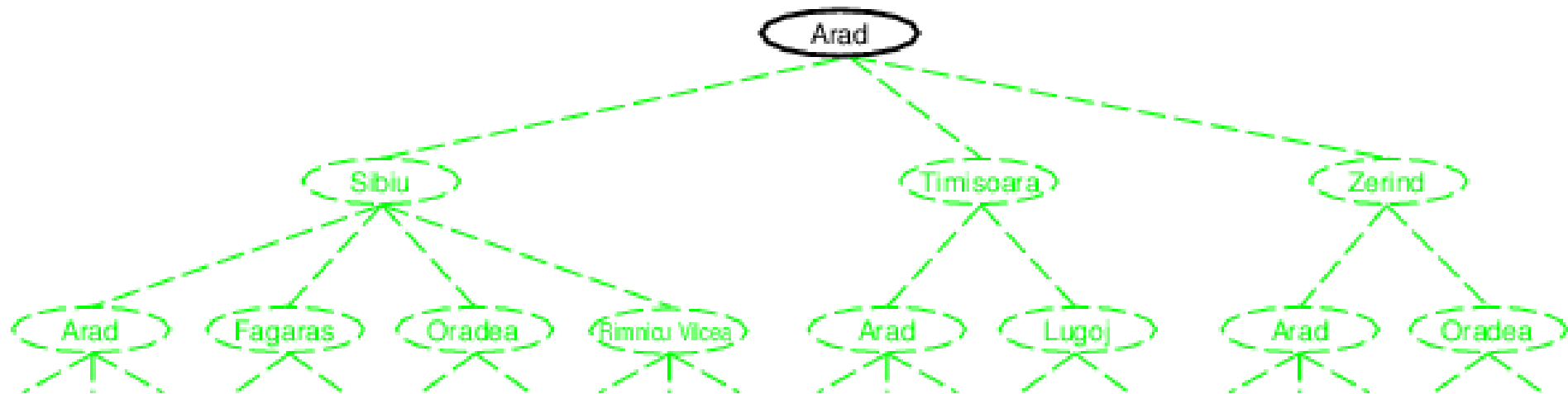
**actions?**: continuous motions of robot joints

**goal test?**: complete assembly

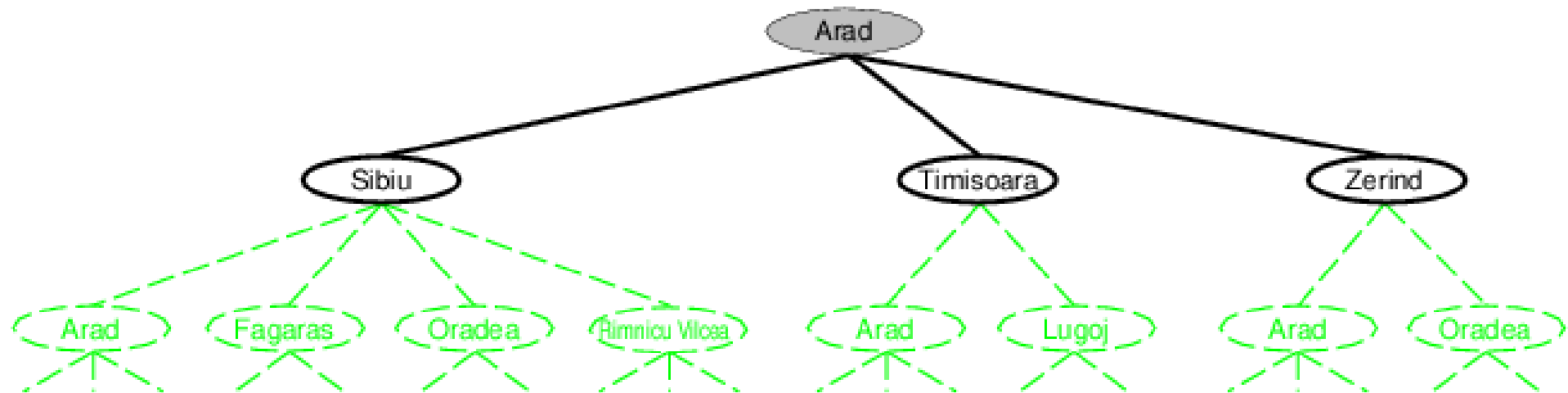
**path cost?**: time to execute

# tree search

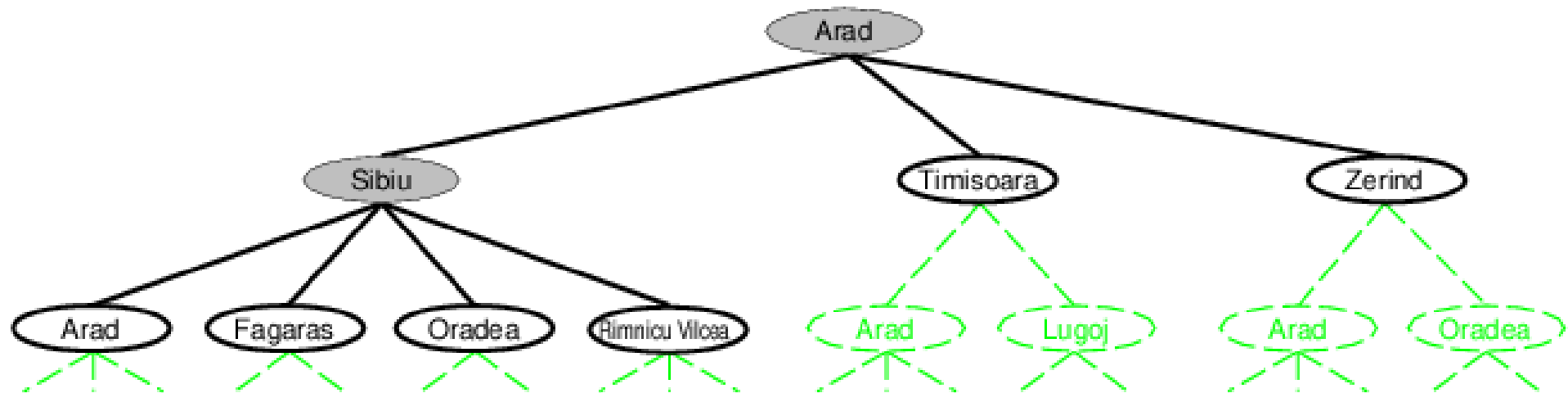
# Tree Search Example



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# Tree Search Example



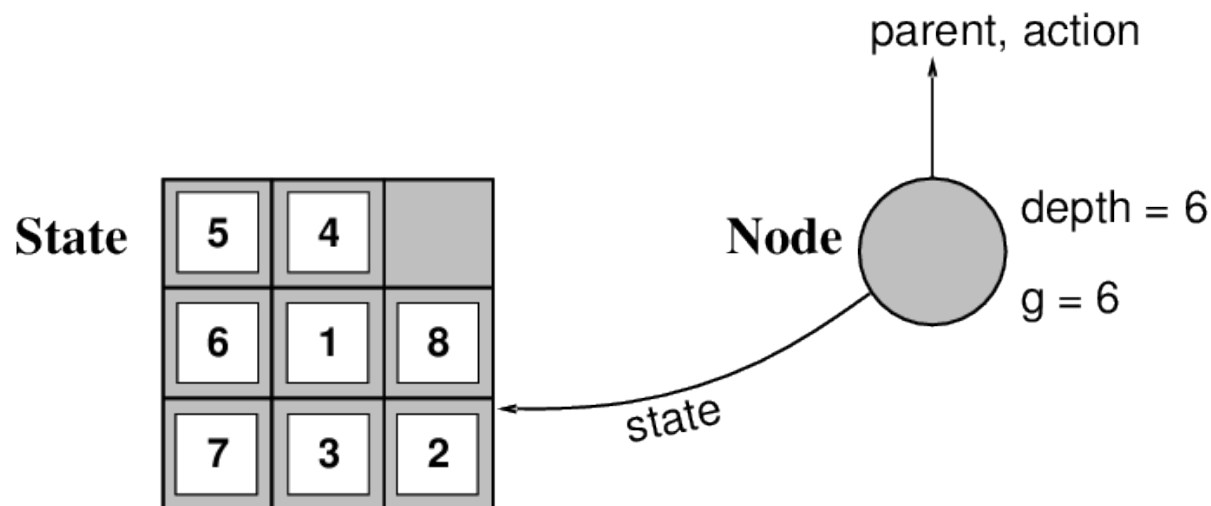
# Tree Search Algorithms

- Basic idea: offline, simulated exploration of state space by generating successors of already-explored states (a.k.a. **expanding** states)

```
function TREE-SEARCH(problem, strategy) returns a solution, or failure
  initialize the search tree using the initial state of problem
  loop do
    if there are no candidates for expansion then return failure
    choose a leaf node for expansion according to strategy
    if the node contains a goal state then return the corresponding solution
    else expand the node and add the resulting nodes to the search tree
  end
```

# Implementation: States vs. Nodes

- A **state** is a (representation of) a physical configuration
- A **node** is a data structure constituting part of a search tree includes **parent**, **children**, **depth**, **path cost**  $g(x)$
- States do not have parents, children, depth, or path cost — nodes do



- A strategy is defined by picking the **order of node expansion**■
- Strategies are evaluated along the following dimensions
  - **completeness**—does it always find a solution if one exists?
  - **time complexity**—number of nodes generated/expanded
  - **space complexity**—maximum number of nodes in memory
  - **optimality**—does it always find a least-cost solution?■
- Time and space complexity are measured in terms of
  - $b$  — maximum branching factor of the search tree
  - $d$  — depth of the least-cost solution
  - $m$  — maximum depth of the state space (may be  $\infty$ )

# Uninformed Search Strategies



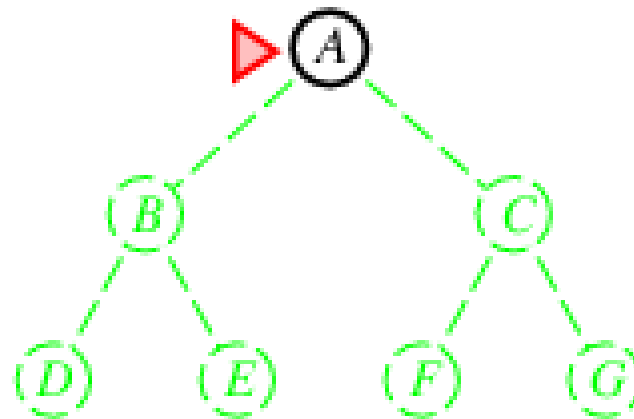
**Uninformed** strategies use only the information available in the problem definition

- Breadth-first search
- Uniform-cost search
- Depth-first search
- Depth-limited search
- Iterative deepening search

# breadth-first search

# Breadth-First Search

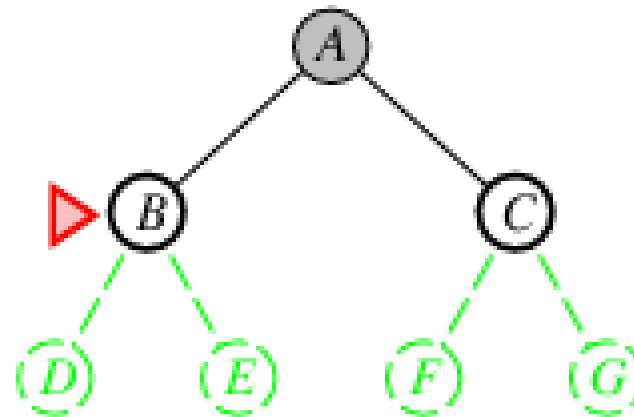
- Expand shallowest unexpanded node
- **Implementation:**  
*fringe* is a FIFO queue, i.e., new successors go at end



*fringe* = (A)

# Breadth-First Search

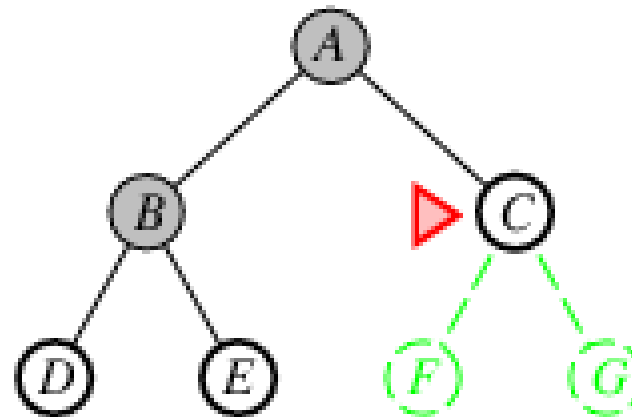
- Expand shallowest unexpanded node
- **Implementation:**  
*fringe* is a FIFO queue, i.e., new successors go at end



*fringe* = (B,C)

# Breadth-First Search

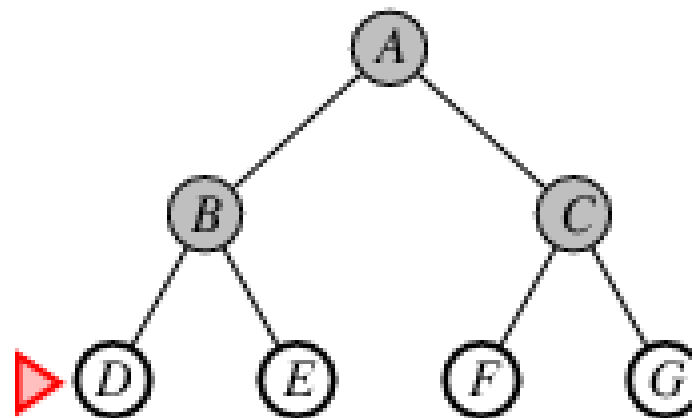
- Expand shallowest unexpanded node
- **Implementation:**  
*fringe* is a FIFO queue, i.e., new successors go at end



*fringe* = (C,D,E)

# Breadth-First Search

- Expand shallowest unexpanded node
- **Implementation:**  
*fringe* is a FIFO queue, i.e., new successors go at end



*fringe* = (D,E,F,G)

# Properties of Breadth-First Search

- **Complete?** Yes (if  $b$  is finite)
- **Time?**  $1 + b + b^2 + b^3 + \dots + b^d + b(b^d - 1) = O(b^{d+1})$ , i.e., exp. in  $d$
- **Space?**  $O(b^{d+1})$  (keeps every node in memory)
- **Optimal?** Yes (if cost = 1 per step); not optimal in general
- **Space** is the big problem

# uniform cost search

# Uniform-Cost Search

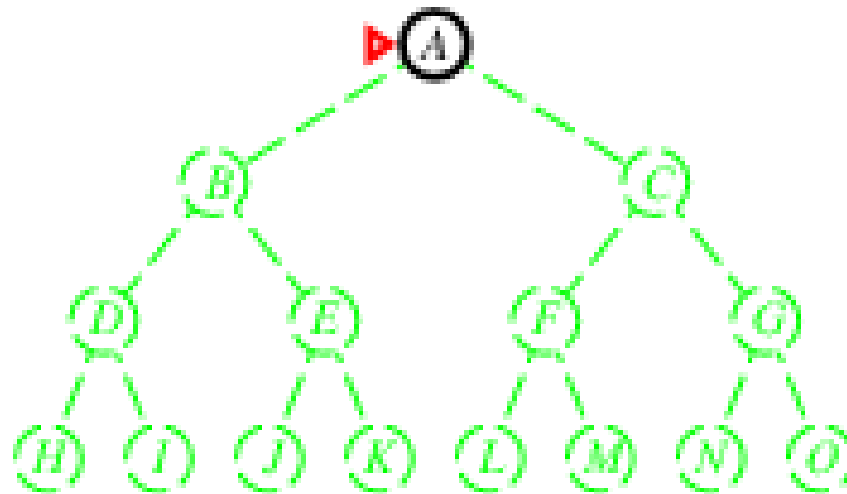


- Expand least-cost unexpanded node
- **Implementation:**  
*fringe* = queue ordered by path cost, lowest first
- Equivalent to breadth-first if step costs all equal
- Properties
  - **Complete?** Yes, if step cost  $\geq \epsilon$
  - **Time?** # of nodes with  $g \leq$  cost of optimal solution,  $O(b^{\lceil C^*/\epsilon \rceil})$   
where  $C^*$  is the cost of the optimal solution
  - **Space?** # of nodes with  $g \leq$  cost of optimal solution,  $O(b^{\lceil C^*/\epsilon \rceil})$
  - **Optimal?** Yes—nodes expanded in increasing order of  $g(n)$

# depth first search

# Depth-First Search

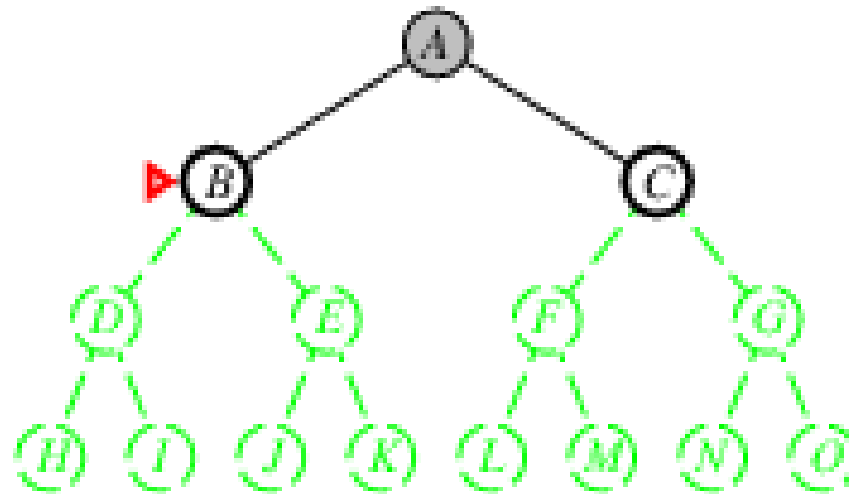
- Expand deepest unexpanded node
- **Implementation:**  
*fringe* = LIFO queue, i.e., put successors at front



*fringe* = (A)

# Depth-First Search

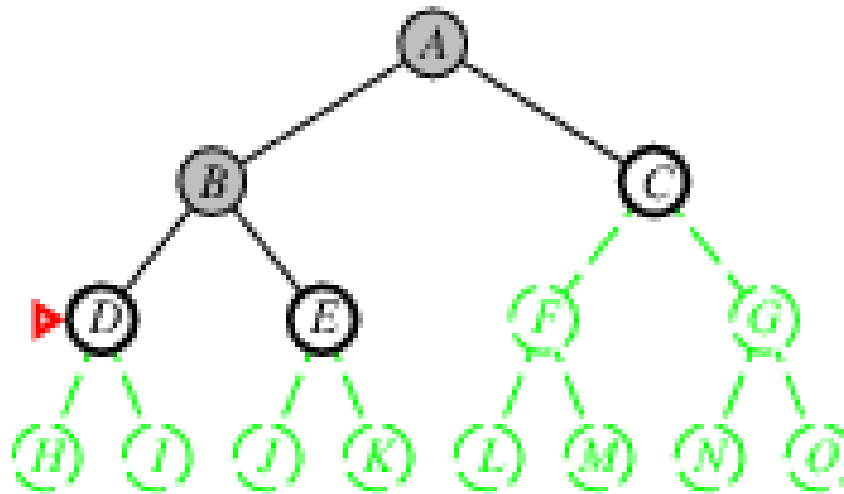
- Expand deepest unexpanded node
- **Implementation:**  
*fringe* = LIFO queue, i.e., put successors at front



*fringe* = (B,C)

# Depth-First Search

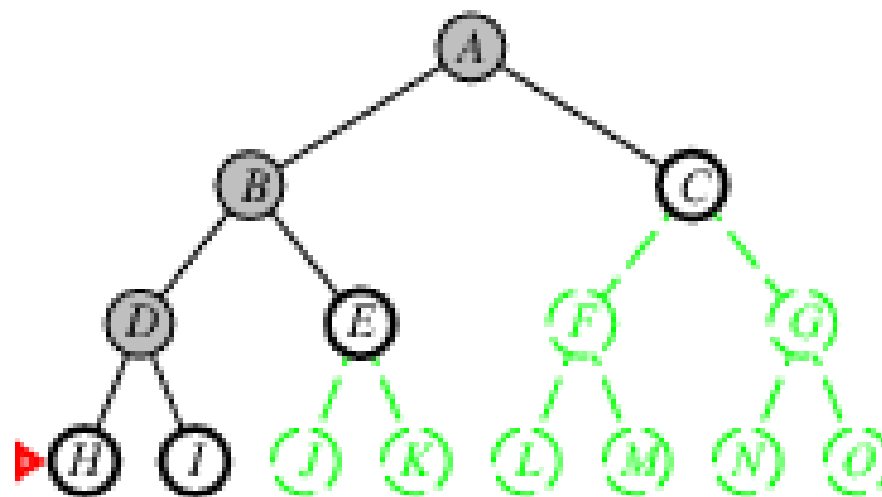
- Expand deepest unexpanded node
- **Implementation:**  
*fringe* = LIFO queue, i.e., put successors at front



*fringe* = (D,E,C)

# Depth-First Search

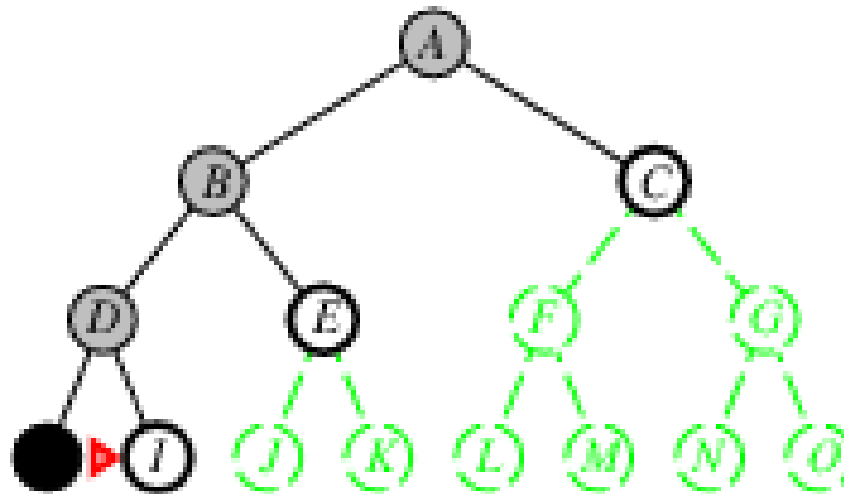
- Expand deepest unexpanded node
- **Implementation:**  
*fringe* = LIFO queue, i.e., put successors at front



*fringe* = (H,I,E,C)

# Depth-First Search

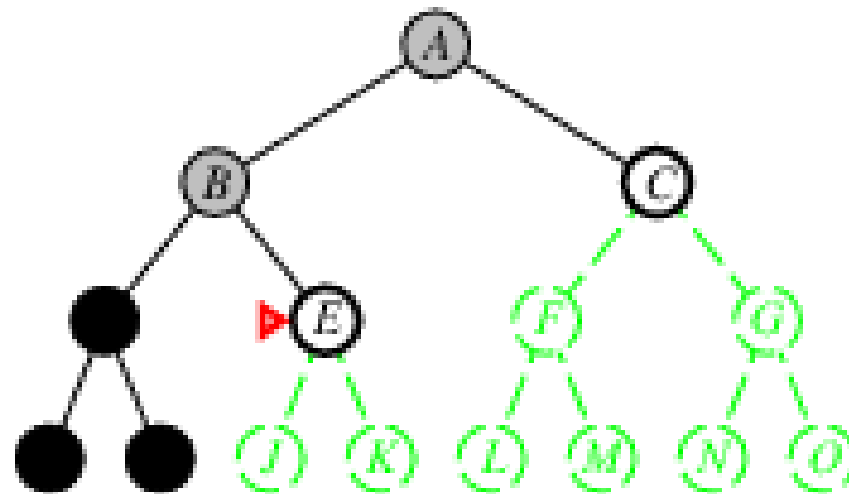
- Expand deepest unexpanded node
- **Implementation:**  
*fringe* = LIFO queue, i.e., put successors at front



*fringe* = (I,E,C)

# Depth-First Search

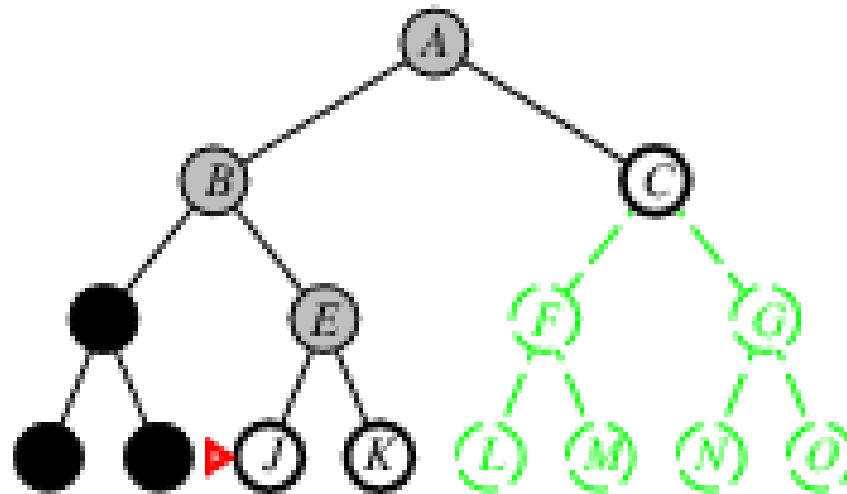
- Expand deepest unexpanded node
- **Implementation:**  
*fringe* = LIFO queue, i.e., put successors at front



*fringe* = (E,C)

# Depth-First Search

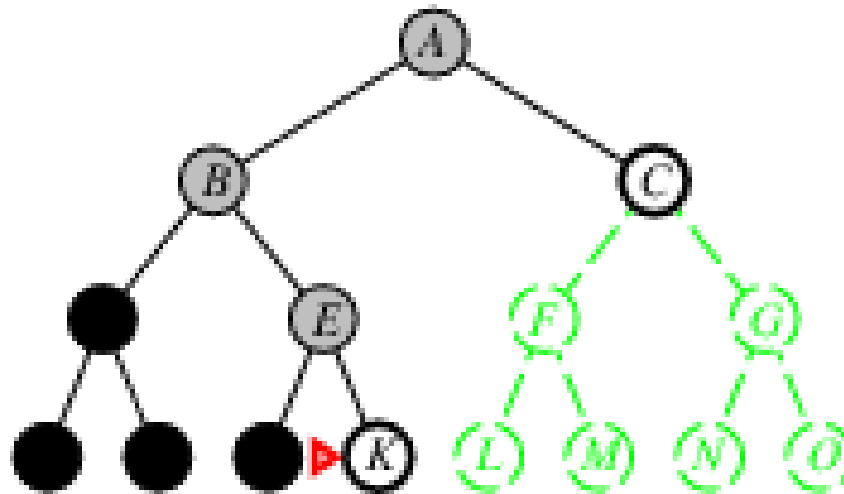
- Expand deepest unexpanded node
- **Implementation:**  
*fringe* = LIFO queue, i.e., put successors at front



*fringe* = (J,K,C)

# Depth-First Search

- Expand deepest unexpanded node
- **Implementation:**  
*fringe* = LIFO queue, i.e., put successors at front

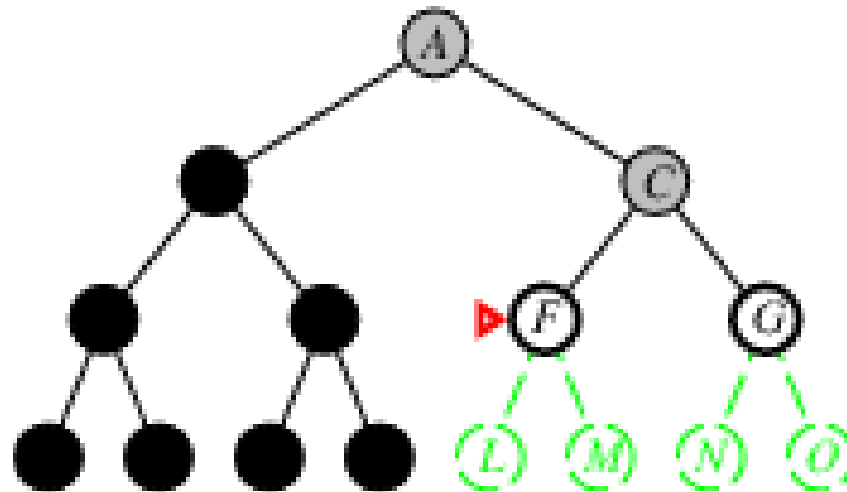


*fringe* = (K,C)



# Depth-First Search

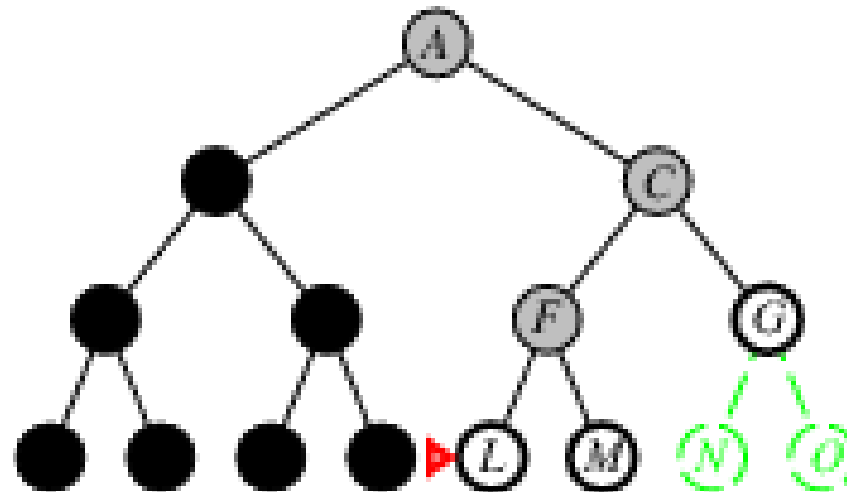
- Expand deepest unexpanded node
- **Implementation:**  
*fringe* = LIFO queue, i.e., put successors at front



*fringe* = (F,G)

# Depth-First Search

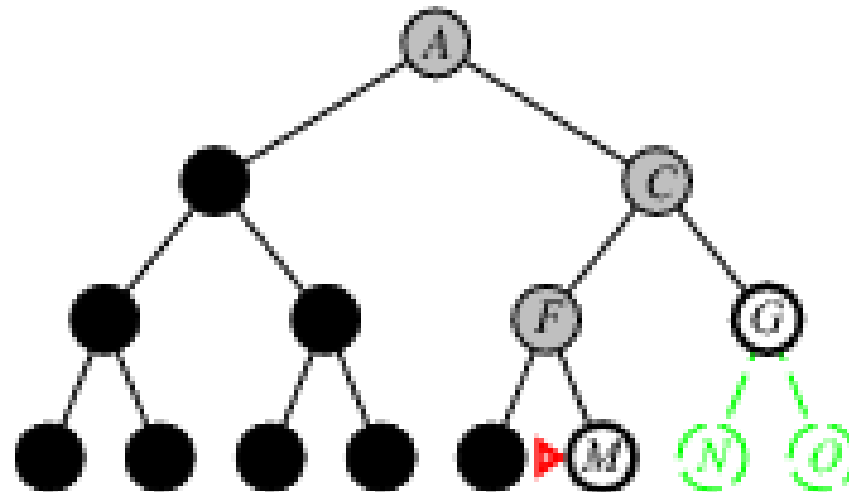
- Expand deepest unexpanded node
- **Implementation:**  
*fringe* = LIFO queue, i.e., put successors at front



*fringe* = (L,M,G)

# Depth-First Search

- Expand deepest unexpanded node
- **Implementation:**  
*fringe* = LIFO queue, i.e., put successors at front



*fringe* = (M,G)

# Properties of Depth-First Search

- **Complete?** ■
  - no: fails in infinite-depth spaces, spaces with loops
  - modify to avoid repeated states along path
    - ⇒ complete in finite spaces■
- **Time?** ■  $O(b^m)$ 
  - terrible if  $m$  is much larger than  $d$
  - but if solutions are dense, may be much faster than breadth-first■
- **Space?** ■  $O(bm)$ , i.e., linear space!■
- **Optimal?** ■ No

# iterative deepening

# Depth-Limited Search



- Depth-first search with depth limit  $l$ , i.e., nodes at depth  $l$  have no successors
- Iterative deepening
  1. Start with depth 1
  2. Carry out depth-first search
  3. Found solution? Terminate.
  4. Otherwise increase depth by 1
  5. Go to step 2

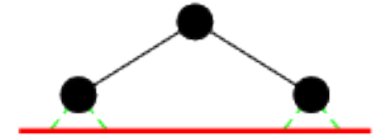
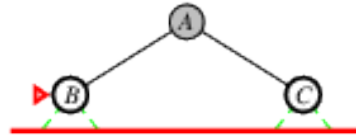
# Iterative Deepening Search $l = 0$

Limit = 0



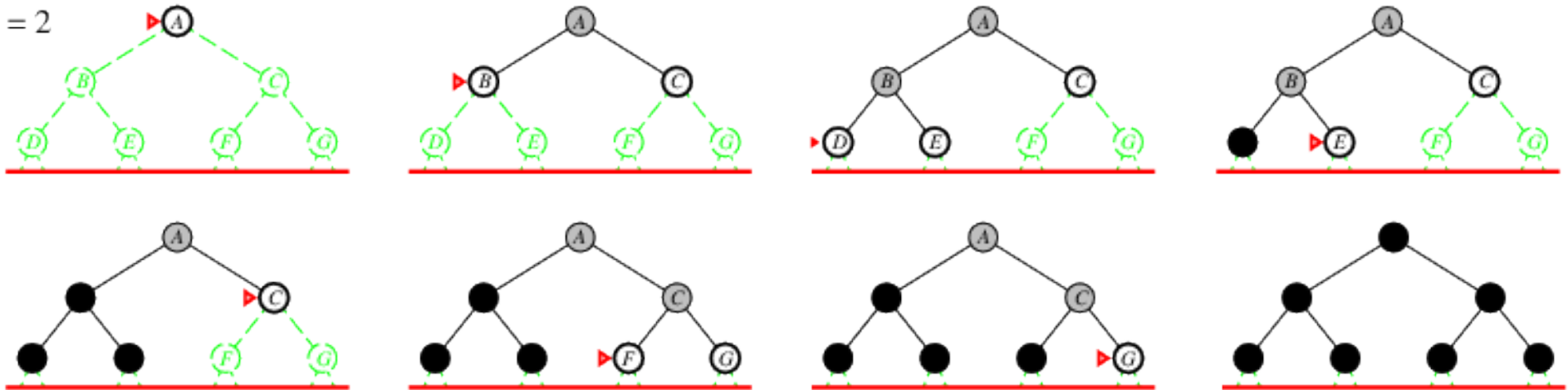
# Iterative Deepening Search $l = 1$

Limit = 1



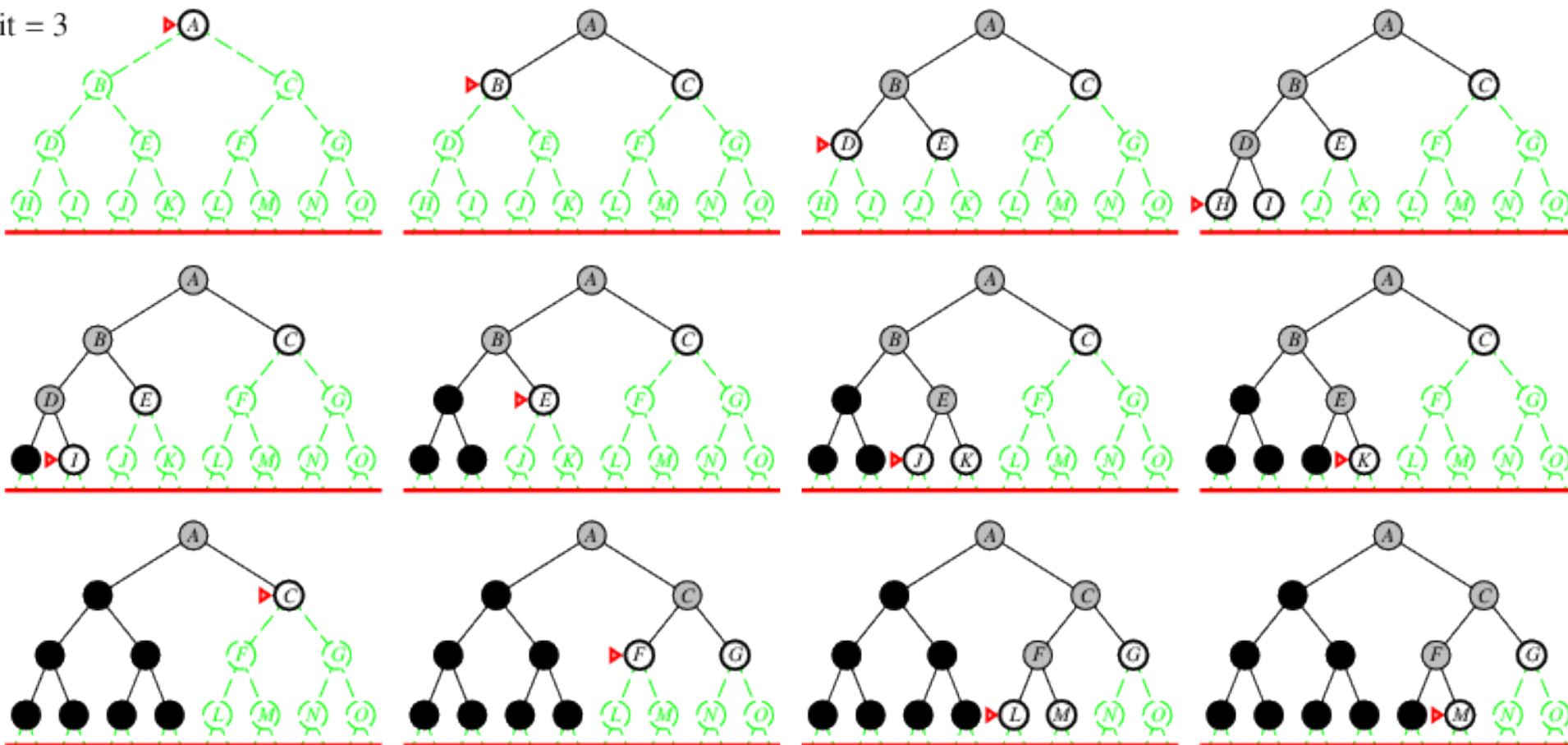
# Iterative Deepening Search $l = 2$

Limit = 2



# Iterative Deepening Search $l = 3$

Limit = 3



# Properties of Iterative Deepening Search



- **Complete?** ■ Yes ■
- **Time?** ■  $(d + 1)b^0 + db^1 + (d - 1)b^2 + \dots + b^d = O(b^d)$  ■
- **Space?** ■  $O(bd)$  ■
- **Optimal?** ■ Yes, if step cost = 1  
Can be modified to explore uniform-cost tree
- Numerical comparison for  $b = 10$  and  $d = 5$ , solution at far right leaf:

$$N(\text{IDS}) = 50 + 400 + 3,000 + 20,000 + 100,000 = 123,450$$

$$N(\text{BFS}) = 10 + 100 + 1,000 + 10,000 + 100,000 + 999,990 = 1,111,100$$

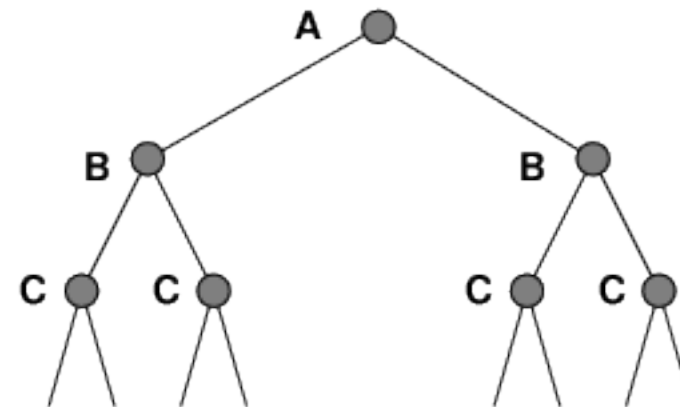
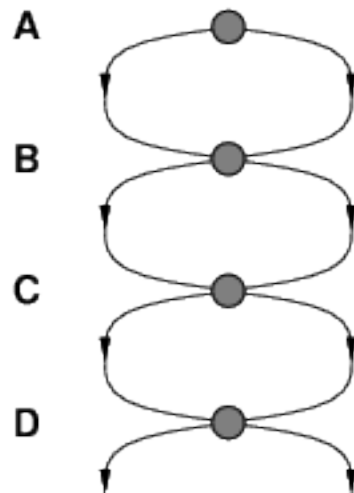
- IDS does better because other nodes at depth  $d$  are not expanded
- BFS can be modified to apply goal test when a node is **generated**

# Summary of Algorithms

Criterion	Breadth-First	Uniform-Cost	Depth-First	Depth-Limited	Iterative Deepening
Complete?	Yes*	Yes*	No	Yes, if $l \geq d$	Yes
Time	$b^{d+1}$	$b^{\lceil C^*/\epsilon \rceil}$	$b^m$	$b^l$	$b^d$
Space	$b^{d+1}$	$b^{\lceil C^*/\epsilon \rceil}$	$bm$	$bl$	$bd$
Optimal?	Yes*	Yes	No	No	Yes*

# Repeated States

Failure to detect repeated states can turn a linear problem into an exponential one



# Summary

- Problem formulation usually requires abstracting away real-world details to define a state space that can feasibly be explored
- Variety of uninformed search strategies
- Iterative deepening search uses only linear space and not much more time than other uninformed algorithms
- Graph search can be exponentially more efficient than tree search